# Documentation: Surya's Text to Math Problem Solver using Gradio

#### **Overview**

This application is an AI-powered **Text to Math Problem Solver** built using **Gradio** and **LangChain**. It takes user-inputted mathematical questions, processes them using **Groq's LLM**, and provides **step-by-step explanations** for the solutions. The system also integrates **Wikipedia search** for additional context and explanations.

#### **Features**

- Accepts math-related text queries.
- Generates numerical expressions and assumptions if needed.
- Provides **step-by-step explanations** for problem-solving.
- Uses **Wikipedia search** for additional knowledge.
- Requires a **GROQ API key** for language model inference.
- Interactive UI using Gradio.
- Offers **real-time response** generation.

## **Technologies Used**

- Python: Core programming language.
- Gradio: Web-based UI for easy interaction.
- LangChain: For handling LLM-based operations and chains.
- **Groq**: AI model inference for generating explanations.
- WikipediaAPIWrapper: Fetches relevant content from Wikipedia.
- **Regular Expressions (re)**: Extracts numerical expressions.
- **dotenv**: Manages API keys securely.

## **Workflow Explanation**

#### 1. Model Initialization

- Loads the **GROQ API key** from user input.
- Initializes **Groq's LLM** (gemma2-9b-it).

#### 2. Tools and Functionalities

#### Wikipedia Search Tool

• Allows searching Wikipedia for relevant math topics.

#### **Math Problem Solver**

- Uses **LLMMathChain** for performing calculations.
- Extracts **numerical expressions** from input queries.
- Assumes **default values for variables** when needed.
- Generates **step-by-step explanations** of the solution.

#### **Reasoning Tool**

- Uses a custom **PromptTemplate** to guide the model.
- Provides **detailed reasoning** behind mathematical solutions.

#### 3. Response Generation

- Combines all tools into an **AI Agent**.
- Processes the question and determines the **best approach**.
- If the model response does not contain steps, it falls back to detailed calculation logic.

## **User Interaction via Gradio**

#### Inputs

- **GROQ API Key** (Required for processing).
- Math Question (The problem statement).

#### **Outputs**

- **Step-by-step solution** explaining the mathematical process.
- Final computed result.
- Assumptions made (if any variables exist in the query).

#### **Buttons & Actions**

• Find my answer → Triggers the AI system to process the input and display results.

## **Error Handling & Edge Cases**

- Invalid API Key → Displays a warning and stops execution.
- Non-Mathematical Inputs → Returns a warning asking for a valid math question.
- No Numerical Expression Found → Provides an error message stating that a valid expression was not detected.
- Unrecognized Variables → Assumes default values and clearly mentions them in the output.

## **Deployment & Execution**

- The **Gradio app** is launched in a **web browser**.
- The share=True parameter allows external users to test the application via a **public Gradio link**.
- Works locally or can be **hosted on cloud platforms**.

## **Use Cases**

- Students & Learners → Step-by-step guidance for solving math problems.
- **Tutors & Teachers** → Automated solution explanations.
- Data Analysts & Researchers → Quick verification of numerical expressions.
- **Developers & Engineers** → AI-powered math problem-solving assistance.

# **Future Improvements**

- Enhanced Equation Parsing → Improve accuracy in extracting and solving complex formulas.
- **Graphical Representations** → Visual solutions with plots/graphs.
- **Voice Input Support** → Allow users to ask math questions verbally.
- Expanded Knowledge Base → Integrate more data sources beyond Wikipedia.

 Multiple AI Models → Use different LLMs for improved reasoning and calculation accuracy.

## **Conclusion**

This **AI-powered math solver** bridges the gap between **text-based queries** and **step-by-step numerical solutions**. With the combination of **Groq LLM**, **Wikipedia search**, **and advanced reasoning models**, it provides a **comprehensive**, **interactive**, **and intelligent problem-solving experience** for users.

## Code:

import gradio as gr
from langchain_groq import ChatGroq
from langchain.chains import LLMMathChain, LLMChain
from langchain.prompts import PromptTemplate
from langchain_community.utilities import WikipediaAPIWrapper
from langchain.agents.agent_types import AgentType
from langchain.agents import Tool, initialize_agent
from dotenv import load_dotenv
import re
import os
#Load environment variables
load_dotenv()
# Initialize the language model
def initialize_model(groq_api_key):
try:

```
llm = ChatGroq(model="gemma2-9b-it", groq_api_key=groq_api_key)
    return llm
 except Exception as e:
    raise Exception(f"Failed to initialize model: {e}")
# Define calculate_with_explanation globally
def calculate_with_explanation(question, llm):
 try:
    math_chain = LLMMathChain.from_llm(llm=llm)
   math_prompt = """
    You are a mathematical assistant. For the given question, provide a numerical expression (no
variables) and a detailed, point-wise explanation of how to solve it. If the question contains variables,
assume reasonable numerical values (e.g., k=1) and state your assumption:
    Question: {question}
   Numerical Expression: <expression>
    Description:
   - Step 1: [First step]
   - Step 2: [Second step]
    - ... [Continue as needed]
    Final Result: <result>
    math_prompt_template = PromptTemplate(input_variables=["question"], template=math_prompt)
    math_explain_chain = LLMChain(llm=llm, prompt=math_prompt_template)
```

```
explanation = math_explain_chain.run({"question": question})
    expr_match = re.search(r'Numerical Expression: (.*?)\n', explanation)
    if not expr_match:
      return f"Question: {question}\nError: No valid numerical expression provided\nDescription:\n-
Step 1: Failed to parse a numerical expression\nFinal Result: N/A"
    expr = expr_match.group(1).strip()
    if re.search(r'[a-zA-Z]', expr):
      assumption = "Assumption: Any variables (e.g., k) set to 1 unless specified."
      expr = re.sub(r'[a-zA-Z]', '1', expr)
      explanation = f'' {explanation}\n{assumption}''
    result = math_chain.run(expr)
    return f" {explanation }\nFinal Result: {result}"
 except Exception as e:
    return f"Question: {question}\nNumerical Expression: {expr if 'expr' in locals() else
'N/A' \nDescription:\n- Error: {e}\nFinal Result: N/A"
# Initialize the Chat Tools
def initialize_tools(llm):
 wikipedia_wrapper = WikipediaAPIWrapper()
 wikipedia_tool = Tool(
    name="Wikipedia",
    func=wikipedia_wrapper.run,
   description="A tool for searching Wikipedia to assist with math problems"
```

```
# Math tool using the global calculate_with_explanation
 calculator = Tool(
    name="Calculator",
    func=lambda q: calculate_with_explanation(q, llm), # Pass llm to the function
    description="Solves math questions with step-by-step explanations."
 # Define the reasoning prompt
 prompt = """
 You are an agent tasked with solving the user's mathematical question. Provide a numerical
expression (no variables) and a detailed, point-wise explanation. If variables are present, assume
reasonable values (e.g., k=1) and note the assumption:
 Question: {question}
 Numerical Expression: <expression>
 Description:
 - Step 1: [First step]
 - Step 2: [Second step]
 - ... [Continue as needed]
 Final Result: <result>
 prompt_template = PromptTemplate(input_variables=["question"], template=prompt)
 try:
```

```
chain = LLMChain(llm=llm, prompt=prompt_template)
    reasoning_tool = Tool(
      name="Reasoning",
      func=chain.run,
      description="Answers math questions with step-by-step explanations."
 except Exception as e:
   raise Exception(f"Reasoning tool setup failed: {e}")
 return [wikipedia_tool, calculator, reasoning_tool]
# Function to generate the response
def generate_response(user_question, groq_api_key):
 try:
    llm = initialize_model(groq_api_key)
    tools = initialize_tools(llm)
   assistant_agent = initialize_agent(
      tools=tools,
      llm=llm,
      agent=AgentType.ZERO_SHOT_REACT_DESCRIPTION,
      verbose=False,
      handle_parsing_errors=True,
    response = assistant_agent.invoke({'input': user_question})
```

```
resp = response['output'] if isinstance(response, dict) and 'output' in response else str(response)
   if "Description:" not in resp:
      return calculate_with_explanation(user_question, llm) # Use global function with llm
   return resp
 except Exception as e:
   return f"Error: {e}"
# Gradio Interface
def gradio_interface(question, api_key):
 if not api_key:
   return "Please provide a GROQ API key."
 if not question or not question.strip():
   return "Please enter a question."
 return generate_response(question, api_key)
# Create Gradio app
with gr.Blocks(title="Surya's Text to Math Problem Solver") as demo:
 gr.Markdown("# Text to Math Problem Solver")
 gr.Markdown("Enter your math question and GROQ API key to get a step-by-step solution!")
 api_key_input = gr.Textbox(label="GROQ API Key", type="password", placeholder="Enter your
GROQ API key here")
 question_input = gr.Textbox(label="Enter your Question", placeholder="e.g., What is 5 + 3 * 2?")
 output = gr.Textbox(label="Response", interactive=False)
 submit button = gr.Button("Find my answer")
```

```
submit_button.click(

fn=gradio_interface,

inputs=[question_input, api_key_input],

outputs=output

)

# Launch the app

demo.launch(share=True)
```